# NOTES 21: CLT-BASED INFERENCE FOR DIFFERENCES

Stat 120 | Fall 2025 Prof Amanda Luby

#### 1 CLT Recap

i	Note

**CLT for 1 Proportion** 

$$\hat{p} \sim N(\underline{\hspace{1cm}},\underline{\hspace{1cm}})$$
 
$$SE_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$$

But, we don't know p, so plug in \_\_\_\_\_ (CI's), \_\_\_\_ (HT's), \_\_\_\_ (conservative)

**i** Note

**CLT for 1 Mean** 

$$\bar{X} \sim N(\underline{\hspace{1cm}},\underline{\hspace{1cm}})$$

where  $SE_{\bar{X}} =$  \_\_\_\_\_. But, we don't know  $\sigma$ , so we plug in s. Then,

$$\frac{\bar{X} - \mu}{\frac{s}{\sqrt{n}}} \sim t_{n-1}$$

**Example** We want to build a 99% confidence interval for  $\mu$ , where  $\bar{X}=290$ , n=120 and s=87.6. The correct form for the CI is:

## 2 CLT for a difference in proportions

Idea:

Population parameter:

Sample statistic:

CLT:

$$\hat{p}_1 - \hat{p}_2 \sim N(\underline{\hspace{1cm}},\underline{\hspace{1cm}})$$

How big is "big enough?"

• 
$$n_1 p_1 >$$
\_\_\_\_\_\_\_  
•  $n_2 p_2 >$ \_\_\_\_\_\_

• 
$$n_2p_2 >$$
\_\_\_\_\_\_

What is the SE?

$$SE_{\hat{p}_1-\hat{p}_2} = \sqrt{\frac{p_1(1-p_1)}{n_1} + \frac{p_2(1-p_2)}{n_2}}$$

but, we don't know  $p_1$  or  $p_2$  . Three options:

- 1. "Plug-in"  $\hat{p}_1$  and  $\hat{p}_2$
- 2. Use  $\hat{p}_{nooled} =$  \_\_\_\_\_\_
- 3. "Conservative" SE: assume  $p = \_\_\_$

**i** Note

**CLT-based confidence interval for a difference in proportions:** 

$$\hat{p}_1 - \hat{p}_2 \pm z^* \times SE_{pluq-in}$$

Note

**CLT-based test for a difference in proportions:** 

$$H_0: p_1 - p_2 = 0 \\$$

$$z = \frac{(\hat{p}_1 - \hat{p}_2) - 0}{SE_{pooled}}$$

## 3 CLT for a difference in means

Idea:

Population parameter:

Sample statistic:

CLT:

$$\bar{x}_1 - \bar{x}_2 \sim N(\mu_1 - \mu_2, SE_{\bar{x}_1 - \bar{x}_2})$$

How big is "big enough?"

Note

Case 1:

• 
$$n_1 >$$
\_\_\_\_\_

• 
$$n_2 >$$
 \_\_\_\_\_

**i** Note

Case 2:

• 
$$n_1 \leq \underline{\hspace{1cm}}$$
 or  $n_2 \leq \underline{\hspace{1cm}}$ 

What is the SE?

$$SE_{\bar{x}_1 - \bar{x}_2} = \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

but, we don't know  $\sigma_1$  or  $\sigma_2,$  so we plug in  $s_1$  and  $s_2$  instead:

$$SE_{\bar{x}_1 - \bar{x}_2} = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

the sampling distribution is no longer a normal distribution, but becomes a \_\_\_\_\_

Note

**CLT-based confidence interval for a difference in means:** 

$$\bar{x}_1 - \bar{x}_2 \pm t_{df}^* \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

**i** Note

CLT-based test for a difference in means:  $H_0: \mu_1 - \mu_2 = 0$ 

$$t_{df} = \frac{(\bar{x}_1 - \bar{x}_2) - 0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

where df = \_\_\_\_\_\_

#### **4 CLT for Matched Pairs**

Idea:

Population parameter:

Sample statistic:	
CLT:	
How big is "big enough"?	
What is the SE?	
i Note	
CLT-based confidence interval for a matched pairs difference:	
i Note	
CLT-based test for a matched pairs difference: $H_0:\mu_d=0$	
$0  \mu a$	